

Forecasting Stock Price Using Machine Learning Technique

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Abstract—Stock market is an emerging sector in any country of the world. Many people directly related to this sector. Stock market prediction is the act of trying to determine the future value of a company stock or other financial instrument. When publicly traded companies issue shares of stock to investors, each of those shares is assigned a monetary value, or price. Stock prices can go up or down depending on different factors. Stock prices can be affected by a number of things including volatility in the market, current economic conditions, and popularity of the company. The successful prediction of a stock's future price could yield significant profit. Along with the development with the stock market, forecasting become an important topic. Since finance market has become more and more competitive, stock price prediction has been a hot research topic in the past few decades. Predicting stock price is regarded a challenging task because stock market is essentially non linear, non-parametric, noisy, and a chaotic system. Trend of a market depends on many things like liquid money human behavior, news related to stock market etc. All this together controls the behavior of trends in a stock market with the advancement of the computing technology we use machine learning technique, like Support Vector Regression, K-nearest-neighbor, Linear Regression, Random forest Regressor, for analyzing time series data to predict stock price. In this paper we try to develop a forecasting model with stacking multiple method to find the best forecast of the stock price.

Keywords— Time Series data, SVR, KNN-Regressor, linear regression, stacking regressor, Random forest regression

I. INTRODUCTION

The goal is to take time series data, find the equation that best fits the data, and be able to forecast out a specific value. Time series data is a continuous data statistical observations recorded over a specific period of time. This model will try to understand the pattern of the continuous data by combining different methods and produce a best fit line that fits the data. The target is to determine the future stock price and improve their strategy for future. Regression models are among the most known regression models used in the machine learning community and recently many researchers have examined their sufficiency in ensembles. Although many methods of ensemble design have been proposed, there is as yet no obvious picture of which method is best. One notable successful adoption of ensemble learning is the distributed scenario. In this work, we propose an efficient distributed method that uses the same training set with the parallel usage of an averaging methodology that combines linear regression and KNN regression models, Support Vector Regression, random Forest Regression. We performed a

comparison of the presented ensemble with other ensembles that use either the linear regression as base learner and the performance of the proposed method was better in most cases. Using averaging methodology, we expect to obtain better results because both theory and experiments show that averaging helps most if the errors in the individual regression models are not positively correlated. Linear regression is a linear approach to modelling the relationship between a scalar response dependent variable and one or more explanatory variables (or independent variables). The case of one explanatory variable is called simple linear regression. For more than one explanatory variable, the process is called multiple linear regression. Support vector machines are supervised learning models with associated learning algorithms that analyze data used for classification and regression analysis. Given a set of training examples, each marked as belonging to one or the other of two categories, an SVM training algorithm builds a model by creating a hyper plane that assigns new examples to one category or the other. In support vector regression (SVR). The model produced by support vector classification (as described above) depends only on a subset of the training data, because the cost function for building the model does not care about training points that lie beyond the margin. In pattern recognition, the k-nearest neighbors algorithm (k-NN) is a non-parametric method used for classification and regression. In both cases, the input consists of the k closest training examples in the feature space. The output depends on whether KNN is used for classification or regression. In case of the kNN regression the output is the property value for the object. This value is the average of the values of its k nearest neighbors. A useful technique can be used to assign weight to the contributions of the neighbors, so that the nearer neighbors contribute more to the average than the more distant ones. A model that combines KNN regression, Linear regression, Support Vector Regression, Random Forest regression model used for predicting stock prices can forecast better price accuracy. The paper is structured as follows. Section II presents the most well-known methods for building ensembles. Section III contains Proposed methodology. Section IV contains Result Analysis. Section V contains conclusion.

II. ENSEMBLES OF REGRESSION MODELS

Bagging is a "bootstrap" ensemble method that creates individuated by randomly drawing, with replacement, N examples - where N is the size of the original set; many of the original examples may be repeated in the resulting training set while others may be left out. After construction of several regression models, averaging the

predictions of each regression model performs the final prediction. Instability (responsiveness to changes in the training data) is a prerequisite for bagging to be effective. Another approach for building ensembles of regression models is to use a variety of learning algorithms on all of the training data and combine their predictions. When multiple regression models are combined using averaging methodology, we expect to obtain good results based on the belief that the majority of experts are more likely to be correct in their decision when they are close in their opinions. Stacked generalization or Stacking, is a more sophisticated approach for combining predictions of different learning algorithms. Stacking combines multiple regression models to induce a higher-level regression model with improved performance. In detail, the original data set constitutes the level zero data and all the base regression models run at this level. The level one data are the outputs of the base regression models. A learning algorithm is then used to determine how the individuals for its ensemble by training each regression model on a random redistribution of the training set. Each regression model's training set is generated by randomly drawing, with replacement, N examples - where N is the size of the original set; many of the accurately learned. After several cycles, the prediction is performed by taking a weighted average of the predictions of each regression model, with the weights being proportional to each regression model's performance on its training set.

III. PROPOSED METHODOLOGY

Generate linear regression from this formula

$$m = (N \sum xy - \sum x \sum y) / (N \sum x^2 - (\sum x)^2) \quad (1)$$

$$b = \sum y - m(\sum x) / N \quad (2)$$

$$y = mx + b \quad (3)$$

knn

$$\text{dist}(A, B) = \sqrt{\sum (x_i - y_i)^2} / 2m$$

$$f(\mathbf{x}, \mathbf{w}) = \sum_{j=1}^m w_j g_j(\mathbf{x}) + b$$

$$L_\epsilon(y, f(\mathbf{x}, \mathbf{w})) = \begin{cases} 0 & \text{if } |y - f(\mathbf{x}, \mathbf{w})| \leq \epsilon \\ |y - f(\mathbf{x}, \mathbf{w})| - \epsilon & \text{otherwise} \end{cases}$$

$$R_{emp}(\mathbf{w}) = \frac{1}{n} \sum_{i=1}^n L_\epsilon(y_i, f(\mathbf{x}_i, \mathbf{w}))$$

Stacking is concerned with combining multiple classifiers generated by different learning algorithms L_1, \dots, L_n on a single data set S , which is composed by a feature vector $S^i = (X^i, Y^i)$.

• The stacking process can be broken into two phases:

1. Generate a set of base-level classifiers C_1, \dots, C_n .

• Where $C_i = L_i(S)$

2. Train a meta-level classifier to combine base level classifier

3. The training set for the meta-level classifier is generated through a leave-one-out cross validation process.

$$\forall i = 1, \dots, n \text{ and } \forall k = 1, \dots, N$$

$$C_k^i = L_k(S - S^i)$$

• The learned classifiers are then used to generate predictions for

$$Y_k^i = C_k^i(X^i)$$

• The meta-level data sets consists of examples of the form where the features are the predictions of the base-level classifiers and the class is the correct class of the example in hand.

A. Abbreviations and Acronyms

Define abbreviations and acronyms the first time they are used in the text, even after they have been defined in the abstract. Abbreviations such as IEEE, SI, MKS, CGS, sc, dc, and rms do not have to be defined. Do not use abbreviations in the title or heads unless they are unavoidable.

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Table Head	Table Column Head		
	Table column subhead	Subhead	Subhead
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a. Sample of a Table footnote. (Table footnote)

Fig. 1. Example of a figure caption. (figure caption)

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[1] G. Eason, B. Noble, and I. N. Sneddon, “On certain integrals of Lipschitz-Hankel type involving products of Bessel functions,” Phil. Trans. Roy. Soc. London, vol. A247, pp. 529–551, April 1955. (references)

[2] J. Clerk Maxwell, A Treatise on Electricity and Magnetism, 3rd ed., vol. 2. Oxford: Clarendon, 1892, pp.68–73.

[3] I. S. Jacobs and C. P. Bean, “Fine particles, thin films and exchange anisotropy,” in Magnetism, vol. III, G. T. Rado and H. Suhl, Eds. New York: Academic, 1963, pp. 271–350.

[4] K. Elissa, “Title of paper if known,” unpublished.

[5] R. Nicole, “Title of paper with only first word capitalized,” J. Name Stand. Abbrev., in press.

[6] Y. Yorozu, M. Hirano, K. Oka, and Y. Tagawa, “Electron spectroscopy studies on magneto-optical media and plastic substrate interface,” IEEE Transl. J. Magn. Japan, vol. 2, pp. 740–741, August 1987 [Digests 9th Annual Conf. Magnetism Japan, p. 301, 1982].

[7] M. Young, The Technical Writer’s Handbook. Mill Valley, CA: University Science, 1989.

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